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# Chi-squared and Associations in Tabular Audiology Data

Muhammad N. Anwar, Michael P. Oakes and Ken McGarry

**Abstract**—In this paper, we have used the chi-squared test and Yule's Q measure to discover associations in tables of patient audiology data. These records are examples of heterogeneous medical records, since they contain audiograms, textual notes and typical relational fields. In our first experiment we used the chi-squared measure to discover associations between the different fields of audiology data such as patient gender and patient age with diagnosis. Then, in our second experiment we used Yule's Q to discover the strength and direction of the significant associations found by the chi-squared measure. We then examined two measures of association commonly used in market basket analysis, support and confidence. These did not yield any further associations. We discuss our findings in the context of producing an audiology decision support system.

**Index Terms**—Audiology, Chi-squared, Confidence, Support, Yule's Q.

## I. INTRODUCTION

Association measures can be used to measure the strength of relationship between the variables in medical data. Discovering associations in medical data has an important role in predicting the patient's risk of certain diseases. Early detection of any disease can save time, money and painful procedures [1]. In our work we are looking for significant associations in heterogeneous audiology data with the ultimate aim of looking for factors influencing which patients would most benefit from being fitted with a hearing aid.

Support and confidence are measures of the interestingness of associations between variables [2, 3]. They show the usefulness and certainty of discovered associations. Strong associations are not always interesting, because support and confidence do not filter out uninteresting associations [4]. Thus, to overcome this problem a correlation measure is augmented to support and confidence. One of the correlation measures popularly used in the medical domain is chi-squared ( $\chi^2$ ).

In section II we describe our database of audiology data. We first use the chi-squared measure to discover significant associations in our data, as described in section III. We then use Yule's Q measure to discover the strength of each of our

significant associations, as described in section IV. In Section V, we find the support and confidence for each of the significant associations, and contrast these with the strengths of the associations found in section III. We draw our conclusions in section VI.

## II. AUDIOLOGY DATA

In this study, we have made use of audiology data collected at the hearing aid out-patient clinic at James Cook University Hospital in Middlesbrough. The data consists of about 180,000 individual records covering about 23,000 audiology patients. The data in the records is heterogeneous, consisting of the following fields:

- 1) Audiograms, which are the graphs of hearing ability at different frequencies (pitches). They consist of two graphs, AC and BC, each obtained for both ears. AC stands for air conduction (which uses sounds from a headphone on the ear for measuring the overall hearing ability), while BC stands for bone conduction (in which sound is given behind the ear at the mastoid bone, to measure the hearing ability of the inner ear – cochlea and auditory nerve). An example of an audiogram for one ear is |80|80|95|95|85|85|20|40|50|65|55|. The first six values are for AC thresholds (the faintest sound the patient can hear in decibels) at 250, 500, 1000, 2000, 4000 and 8000 Hz, and the last five values are for BC thresholds at the same frequencies excluding 8000Hz.
- 2) Structured data: gender, date of birth, diagnosis and hearing aid type, as stored in a typical database, e.g. |M|, |09-05-1958|, |TINNITUS|, |BE18|.
- 3) Textual notes: specific observations made about each patient, such as |HEARING TODAY NEAR NORMAL - USE AID ONLY IF NECESSARY|.

In general, these audiology records represent all types of medical records because they involve both structured and unstructured data.

## III. DISCOVERY OF ASSOCIATIONS WITH THE CHI-SQUARED TEST

The Chi-squared test is a simple way to provide estimates of quantities of interest and related confidence intervals [5]. It is a measure of associations between variables (such as the fields of the tables in a relational database) where the variables are nominal and related to each other [6]. The Chi-squared test is popular in the medical domain because of its simplicity. It has been used in pharmacology to classify text according to subtopics [7]. The resulting chi-squared

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value is a measure of the differences between a set of observed and expected frequencies within a population, and is given by the formula:

$$X^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}} \quad [5]$$

where r is the number of unique terms in a particular field of the patient records such as diagnosis or hearing aid type, corresponding to rows in Table 1. c is the number of categories in the data (such as age or gender) corresponding to columns in Table 1.

Table 1  
Observed and Expected frequencies for diagnosis

Diagnosis	Age≤54	Age>54	R. Total
DOWNS	12 (6.18) [33.92]	0 (5.82) [33.92]	12
FAM	11 (5.66) [28.50]	0 (5.34) [28.50]	11
FAMILIAL	18 (9.26) [76.32]	0 (8.74) [76.32]	18
INFO	4 (11.32) [53.62]	18 (10.68) [53.62]	22
REV	9 (8.75) [0.06]	8 (8.25) [0.06]	17
TINNITUS	535 (580.02) [2027.24]	592 (546.98) [2027.24]	1127
OTHERS	113 (80.80) [1036.71]	44 (76.20) [1036.71]	157
C. Total	702	662	1364

R. Total stands for row total

C. Total stands for column total

Expected frequencies are in ( )

(Observed frequency – Expected frequency)<sup>2</sup> are in [ ]

Table 1 is produced for 7 diagnoses occurring in the hearing diagnosis field. For example, if 535 of the hearing diagnosis fields of the records of patients ‘aged ≤ 54’ years contained the diagnosis ‘TINNITUS’, we would record a value of 535 for that term being associated with that category. These values were the “observed” values, denoted  $O_{ij}$  in the formula above. The corresponding “expected” values  $E_{ij}$  were found by the formula:

Row total x Column total / Grand Total

The row total for ‘TINNITUS’ diagnosis is the total number of times the ‘TINNITUS’ diagnosis was assigned to patients in both age categories = 535 + 592 = 1127. The column total for ‘age≤54’ is the total number of patients in that age group over all 7 diagnoses = 702. The grand total is the total number of patient records in the study = 1364. Thus the “expected” number of patients diagnosed with ‘TINNITUS’ in the ‘age≤54’ group was 1127 \* 702 / 1364 = 580.02. The significance of this is that the expected value is greater than

the observed value, suggesting that there is a negative degree of association between the ‘TINNITUS’ diagnosis and the category ‘age≤54’. The remainder of the test is then performed to discover if this association is statistically significant.

Next the  $O_{ij}$  and  $E_{ij}$  values were used to calculate an overall chi-squared value for the relationship between each of the text variables (hearing aid type, mould, mask and text comments fields) and age, as shown in Table 2. From this data we can show, with 99.9% confidence, that these keywords were not randomly distributed, and that some keywords definitely are associated with age. Similarly the associations of each of these variables with gender are shown in Table 3. Here we see that there are significant associations between the comments text, hearing aid type and mould with gender, but there are no significant associations between diagnosis and mask with gender.

Table 2  
Overall  $\chi^2$  with age

Fields	Overall $\chi^2$	Degrees of freedom (df)	P
Comments text	4624.99	851	P < 0.001
Diagnosis	82.07	6	P < 0.001
Hearing aid type	750.12	46	P < 0.001
Mask	15.15	3	P < 0.001
Mould	342.68	18	P < 0.001

Table 3  
Overall  $\chi^2$  with gender

Fields	Overall $\chi^2$	Degrees of freedom (df)	P
Comments text	2042.51	910	P < 0.001
Diagnosis	6.31	6	P = 0.392
Hearing aid type	729.10	49	P < 0.001
Mask	4.17	3	P = 0.243
Mould	288.79	17	P < 0.001

Having shown that overall, some keywords are more associated with some category; the next step was to discover exactly which individual keywords were most (and least) associated with each category. To do this, we considered the individual contributions of each word in each category to the overall chi-squared value for each text field, found by the formula

$$X^2 = \frac{(O_{ij} - E_{ij})^2}{E_{ij}}$$

for each word in each category. To use the chi-squared test the expected frequency values must be all at least 1, and most should exceed 5 [8]. To be on the safe side, we insisted that for each word, all the expected values should be at least 5, so all words failing this test were grouped into a single class called “OTHERS”. Since we were in effect performing many individual statistical tests, it was necessary to use the

Bonferroni correction [5] to control the rate of Type I errors where a word spuriously appears to be typical of a cluster.

Table 4  
Categories with positive and negative keywords for age

	Positive keywords	negative keywords
Age≤54	** OTHERS	** <i>Not found</i>
Age>54	** <i>Not found</i>	** OTHERS, FAMILIAL
Age≤56	**** <i>Not found</i>	**** <i>Not found</i>
Age>56	**** <i>Not found</i>	**** <i>Not found</i>
Age≤70	* masker, tinnitus, 2000, help, progress, dna, fta, counsel  *** PPCL, ITEHH, ITENL, -, PFPPCL, PPC2, BE101, OTHERS  ***** N8, SIL, V2	* dv, staff, map, ref, wax, reqd, gp, contact, cic, insert, reinstruct  *** ITENN, BE34, ITENH, BE36  ***** 2107V1
Age>70	* dv, staff, map, ref, wax, reqd, gp, contact, cic, insert, reinstruct  *** ITENN, BE34, ITENH, BE36  ***** 2107V1	* masker, tinnitus, 2000, help, progress, dna, fta, counsel  *** PPCL, ITEHH, ITENL, -, PFPPCL, PPC2, BE101, OTHERS  ***** N8, SIL, V2

Note: words after \* are for comment text  
\*\* are for diagnosis  
\*\*\* are for hearing aid type  
\*\*\*\* are for mask  
\*\*\*\*\* are for mould

We wished to be 99.9% confident that a particular keyword was typical of a particular cluster, the corresponding significance level of 0.001 had to be divided by the number of simultaneous tests, i.e. the number of unique words times the number of categories. In the case of words in the text fields,

this gave a corrected significance level of  $0.001 / (7 * 2) = .0000714286$ . Using West's chi-squared calculator [9], for significance at the 0.001 level with one degree of freedom, we obtained a chi-squared threshold of 15.77. Thus each word in each category with an individual contribution to the overall chi-squared value of more than 15.77 was taken to be significantly associated with that category at the 0.001 level.

Table 5  
Categories with positive and negative keywords for gender

	positive keywords	negative keywords
Male	* he, wife  ** <i>Not found</i>  *** ITEHH, ITENH  **** <i>Not found</i>  ***** V2, N8, IROS	* dv  ** <i>Not found</i>  *** ITEHN, BE34, ITENN  **** <i>Not found</i>  ***** <i>Not found</i>
Female	* dv  ** <i>Not found</i>  *** ITEHN, BE34, ITENN  **** <i>Not found</i>  ***** <i>Not found</i>	* he, wife  ** <i>Not found</i>  *** ITEHH, ITENH  **** <i>Not found</i>  ***** V2, N8, IROS

Note: words after \* are for comment text  
\*\* are for diagnosis  
\*\*\* are for hearing aid type  
\*\*\*\* are for mask  
\*\*\*\*\* are for mould

Words associated with categories with 95% confidence were deemed typical of those categories if  $O > E$ , otherwise they were deemed atypical. The words most typical and atypical of our two categories, age and gender, are shown in Tables 4 and 5. The discovered associations seem intuitively reasonable. For example, it appears that the patients with 'age>70' required domestic visits (DV) and had the problem of wax. The words tinnitus (ringing in the ears) and masker (a machine for producing white noise to drown out tinnitus) were atypical of this category. The hearing aid types

associated with this category were those with high gain, while low hearing aid types were negatively associated with this category.

For these experiments, we used all the records available in the database for each field under study, keeping the criterion that none of the field values should be empty. In Table 4, values of 54 and 56 were calculated as the median ages of the diagnosis and mask records respectively. 70 was the median age of the records for comment text, hearing aid type and mould. In both tables (Table 4 and Table 5) some keywords in the comments text were abbreviations such as 'DV' for 'Domestic Visit' and 'DNA' for 'Did Not Attend'. 'Tinnitus' appears as 'tinnitu' in the tables, since all the text was passed through Porter's stemmer [10] for the removal of grammatical endings. Similarly 'unabl' is the stemmed form of 'unable'.

#### IV. MEASURES OF ASSOCIATION IN CATEGORICAL DATA

Yule's Q is a measure to find the strength of association for between categorical variables. Unlike the chi-squared test, which tells us how certain we can be that a relationship between two variables exists, Yule's Q gives both the strength and direction of that relationship [6]. In the following 2 x 2 table,

	Present	Absent
Present	A	B
Absent	C	D

Yule's Q is given by

$$Q = \frac{AD - BC}{AD + BC} \quad [2]$$

where A, B, C and D are the observed quantities in each cell. Yule's Q is in the range -1 to +1, where the sign indicates the direction of the relationship and the absolute value indicates the strength of the relationship. Yule's Q does not distinguish complete associations (where one of the cell values = 0) and absolute relationships (where two diagonally opposite cell values are both zero), and is only suitable for 2 x 2 tables.

In Tables 6 – 10, Yule's Q values for age with comment text, diagnosis, hearing aid type, mask and mould are given. Similarly, in the Table 11 – 13, Yule's Q values for gender with comment text, hearing aid type and mould are given. "(P)" and "(A)", stand for present and absent.

In Table 6, a Yule's Q value of 0.75 shows that there is a positive association between the keyword 'progress' and the category 'age<=70', which can be restated as a negative association between the keyword 'progress' and the category 'age>70'. In Table 7, for 'diagnosis' there is an absolute association between 'FAMILIAL' and 'age<=54', resulting in a Yule's Q value of 1. This should be viewed in comparison to the chi-squared value for the same association, 17.20 (p < 0.001), showing both that the association is very strong and that we can be highly confident that it exists. The presence of this association shows that a higher proportion of younger people report to the hearing aid clinic with familial (inherited) deafness than older people.

Table 6  
Yule's Q for comment text and age

Comment text	age <= 70 (P)	age >70 (P)	age<=70 (A)	age>70 (A)	Yule's Q
progress	93	13	46833	45555	0.75
dna	105	20	46821	45548	0.67
masker	565	126	46361	45442	0.63
tinnitus	385	123	46541	45445	0.51
help	222	84	46704	45484	0.44
counsel	191	80	46735	45488	0.40
2000	288	125	46638	45443	0.38
fta	542	332	46384	45236	0.23
gp	370	615	46162	55060	-0.16
wax	341	601	46191	55074	-0.19
ref	248	487	46284	55188	-0.24
contact	37	129	46495	55546	-0.49
insert	23	102	46509	55573	-0.58
reqd	15	111	46517	55564	-0.72
cic	10	76	46522	55599	-0.73
staff	17	132	46515	55543	-0.73
map	15	125	46517	55550	-0.75
dv	29	245	46503	55430	-0.75
reinstruct	8	68	46524	55607	-0.75

Table 7  
Yule's Q for diagnosis and age

Diagnosis	age <= 54 (P)	age >54 (P)	age<=54 (A)	age>54 (A)	Yule's Q
FAMILIAL	18	0	684	662	1.00
OTHERS	113	44	589	618	0.46

Table 8  
Yule's Q for hearing aid type and age

Hearing aid type	age<= 70 (P)	age> 70 (P)	age<= 70 (A)	age>70 (A)	Yule's Q
PFPPCL	42	1	11105	10899	0.95
PPCL	78	5	11069	10895	0.88
BE101	44	4	11103	10896	0.83
PPC2	53	6	11094	10894	0.79
ITENL	123	35	11024	10865	0.55
OTHERS	103	37	11044	10863	0.46
ITEHH	536	317	10611	10583	0.26
-	4668	3947	6479	6953	0.12
BE34	640	882	10507	10018	-0.18
ITENH	403	592	10744	10308	-0.21
ITENN	683	1063	10464	9837	-0.25
BE36	97	203	11050	10697	-0.37

Table 9  
Yule's Q for mask and age

Mask	age ≤ 56 (P)	age >56 (P)	age≤56 (A)	age>56 (A)	Yule's Q
AMTI	50	25	257	261	0.34
ITEM	20	12	287	274	0.23
OTHERS	230	230	77	56	-0.16
RM675	7	19	300	267	-0.51

Table 10  
Yule's Q for mould and age

Mould	age ≤ 70 (P)	age >70 (P)	age≤70 (A)	age>70 (A)	Yule's Q
N8	261	94	10873	10805	0.47
SIL	255	101	10879	10798	0.43
V2	575	397	10559	10502	0.18
2107V1	601	913	10533	9986	-0.23

Table 11  
Yule's Q for comment text and gender

Comment text	M (P)	F (P)	M (A)	F (A)	Yule's Q
he	67	2	46465	55673	0.95
wife	44	2	46488	55673	0.93
dv	80	254	46452	55421	-0.45

Table 12  
Yule's Q for hearing aid type and gender

Hearing aid type	M (P)	F (P)	M (A)	F (A)	Yule's Q
ITEHH	665	201	11080	12467	0.58
ITENH	725	295	11020	12373	0.47
ITEHN	1280	1732	10465	10936	-0.13
ITENN	734	1038	11011	11630	-0.14

Table 13  
Yule's Q for mould and gender

Mould	M (P)	F (P)	M (A)	F (A)	Yule's Q
IROS	80	24	11671	12644	0.57
V2	640	342	11111	12326	0.35
N8	253	141	11498	12527	0.32

Familial deafness is relatively rare but can affect any age group, while "others" would include "old-age deafness" (presbycusis) which is relatively common, but obviously restricted to older patients. However, in Table 10, Yule's Q for 'V2' is 0.18, which shows only a weak association between mould and 'age≤70', while the chi-squared value for the same association of 30.25 ( $P < 0.001$ ), showed that it is highly likely that the association exists. In Table 12, Yule's Q for 'ITEHN' (a type of hearing aid worn inside the ear) is

-0.13, which shows a weak negative association between 'ITEHN' and 'male', or in other words, a weak positive association between 'ITEHN' and 'female'. In comparison, the chi-squared value given in Table 5 for the same association of 43.36 ( $P < 0.001$ ), showed that we can be highly confident that the relationship exists. These results show the complementary nature of the chi-squared and Yule's Q results: in all three cases the chi-squared value was highly significant, suggesting that the relationship was highly likely to exist, while Yule's Q showed the strength (strong in the first case, weak in the others) and the direction (positive in the first two cases, negative in the third) of the relationship differed among the three cases.

## V. SUPPORT AND CONFIDENCE FOR ASSOCIATIONS

In market basket analysis,  $A$  might refer to transactions in a travel agent's where a holiday is booked and  $B$  might refer to transactions in a travel agent's where money is changed. Support is the frequency of transactions that contain both the attributes ( $A \cup B$ ) of the rule  $A \Rightarrow B$  [11]. It can also be given as a probability  $P(A \cup B)$  [4]. Confidence is the frequency of transactions containing  $A$  that also contain  $B$  in the rule  $A \Rightarrow B$  [11]. It is also given as probability  $P(B/A)$  [4]. Confidence is also defined as the ratio of  $\text{support}(A \cup B)/\text{support}(A)$  [12]. It is a measure of the strength of a rule. To summarize,

$$\text{support}(A \Rightarrow B) = P(A \cup B)$$

$$\text{confidence}(A \Rightarrow B) = P(B/A) = \frac{\text{support}(A \cup B)}{\text{support}(A)}$$

[4]

Discovery of rules in the form  $A \Rightarrow B$  would help in the development of an audiological decision support system. In general,  $A$  would be easy to determine variables such as age and gender, while  $B$  would be the goals of the system, such as choice of hearing aid-type, masker, mould and diagnosis.

We calculated support and confidence for all keywords (excluding those found only in the textual comments field). In Tables 14 and 15, support and confidence values are calculated for diagnosis with age and gender respectively. In both tables, all keywords were taken that produced 'expected frequencies  $\geq 5$ ' in the chi-squared analysis, all other words being included in the 'OTHERS' category. In Table 14, support for 'age≤54'  $\Rightarrow$  'TINNITUS' was calculated as  $535/1364 = 0.39$ , since (as shown in Table 1) there were 535 patients with 'age ≤ 54' who were diagnosed with tinnitus (the observed frequency) and 1364 records altogether where a diagnosis was given (grand total). Similarly, in Table 15, confidence for 'age≤54'  $\Rightarrow$  'TINNITUS' was calculated as  $535/702$  where 535 is the observed frequency and 702 is the sum of all observed frequencies for 'age≤54' (column total). Unfortunately, in our data set, the support and confidence was very weak for all relations except for those involving the diagnosis 'TINNITUS', although we know (from section III) that these associations were not found to be significant at  $p < 0.05$  by the chi-squared test. The main reason for the lack of findings using support and confidence was that in the vast majority of cases where a diagnosis was given, it was

'TINNITUS', thus swamping the data for all the other diagnoses. This is probably because in many cases, the task of the hearing aid clinic is to treat tinnitus, and thus it is important to record the diagnosis. The other diagnoses are not treated directly in the hearing aid clinic, although they do influence the shape of the audiogram which is the most important criterion in hearing aid selection.

Table 14  
Support and Confidence for diagnosis and age

	age≤54		age>54	
Diagnosis	Supp	Conf	Supp	Conf
DOWNES	0.01	0.02	0.00	0.00
FAM	0.01	0.02	0.00	0.00
FAMILIAL	0.01	0.03	0.00	0.00
INFO	0.00	0.01	0.01	0.03
REV	0.01	0.01	0.01	0.01
TINNITUS	0.39	0.76	0.43	0.89
OTHERS	0.08	0.16	0.03	0.07

Note: 'Supp' denotes support  
'Conf' denotes confidence

Table 15  
Support and Confidence for diagnosis and gender

	Male		Female	
Diagnosis	Supp	Conf	Supp	Conf
DOWNES	0.01	0.01	0.01	0.02
FAM	0.00	0.01	0.00	0.01
FAMILIAL	0.00	0.01	0.01	0.02
INFO	0.01	0.01	0.01	0.02
MENINGITIS	0.00	0.01	0.00	0.00
REV	0.01	0.01	0.01	0.01
TINNITUS	0.44	0.82	0.37	0.81
OTHERS	0.07	0.12	0.05	0.11

Note: 'Supp' denotes support  
'Conf' denotes confidence

## VI. CONCLUSION

In this work we looked for significant associations in heterogeneous audiology data as part of an overall project with the ultimate aim of looking for factors influencing which patients would most benefit most from being fitted with a hearing aid. We have discovered typical and atypical words related to different fields of audiology data, by first using the chi-squared measure to show which relations most probably exist, then using Yule's Q measure of association to find the strength and direction of those relations. We also calculated support and confidence for all relations between age and diagnosis, and gender and diagnosis, but were unable to find many rules with high support and confidence due to the very high proportion of one type of diagnosis ('TINNITUS') in the records. However, we feel that given an audiology database where a diagnosis was routinely recorded for every patient, more rules in the form  $A \Rightarrow B$  would be found. This would form the basis of a decision support system for audiologists,

where the variables in  $A$  would be easily obtained data such as age, gender, and audiogram, and  $B$  would be predictions of the most suitable hearing aid, tinnitus, masker or diagnosis.

It may be that for the 2 x 2 contingency tables the odds ratio would be a better measure of effect rather than support and confidence. This will be examined in future work.

Rules found by data mining should not only be accurate and comprehensible, but also "surprising". McGarry presents a taxonomy of "interestingness" measures whereby the value of discovered rules may be evaluated [13]. In this paper we have looked at objective interestingness criteria, such as statistical significance and the confidence and support for the discovered rules, but we have not yet considered subjective criteria such as unexpectedness and novelty. These require comparing machine-derived rules with the prior expectations of domain experts. A very important subjective criterion is "actionability", which includes such considerations as impact: will the discovered rule lead to any changes in current audiological practice?

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